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### Alternative Splicing from RNA-seq Data without the Genome

8th Special Interest Group meeting on Alternative Splicing AS-SIG, ISMB/ECCB 2011
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### Motivations and Challenges

# Detecting Alternative Splicing (AS) variations from RNA-seq data

- No specific tools for large-scale inference of AS variations among gene transcripts
- Our goal: identification of AS variations without a reference genome

### Motivations and Challenges

- Reference genome is not always available
- RNA-seq data alignment against the genome is too expensive

#### Our Solution

fast\* construction of a graph representation of AS variations from RNA-seq data without a reference genome

\*linear time w.r.t. the number of reads



### RNA-seq Data

- Basic Features:
  - Short sequences (30 400bp)
  - Depth sequencing → Millions / Billions of sequences
  - Quality
  - Unknown error
- High-throughput sequencing platforms:
  - SOLiD, Illumina, Roche's 454, HeliScope

# RNA-seq Analysis: State of Art

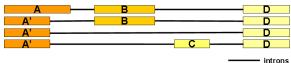
- Read Mapping (Spliced Aligners)
  - Exon-first methods (MapSplice, SpliceMap, Tophat)
  - Seed-extend methods (GSNAP, QPALMA)
- Expression Quantification
  - Gene quantification (Alexa-seq ,ERANGE, NEUMA)
  - Isoform quantification (Cufflinks, MISO, RSEM)
- Transcriptome Reconstruction
  - Genome-guided assembly (Scripture, Cufflinks)
  - Genome-independent assembly (Velvet, TransABySS, Trinity)

NATURE METHODS, June 2011



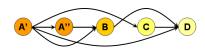
# Our Goal: Isoform Graph

Gene isoforms



Set of blocks

Isoform graph



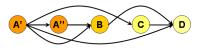
### Our Goal: Isoform Graph

#### **Isoform Graph Reconstruction**

**Input**: a set of RNA-seq reads from unknown gene transcripts



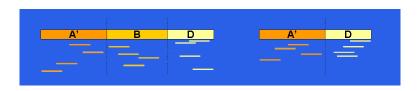
Isoform Graph:



# Our Approach

#### **Isoform Graph Reconstruction**

**Input**: a set of RNA-seq reads from unknown gene transcripts



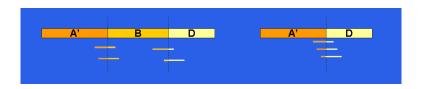
Unspliced reads:



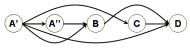
# Our Approach

#### **Isoform Graph Reconstruction**

**Input**: a set of RNA-seq reads from unknown gene transcripts



• Spliced reads:

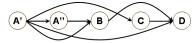


### Method Outline

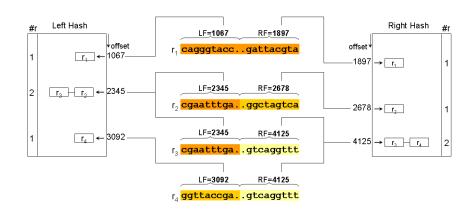
- Hashing input reads
  - Input set partitioning → Unspliced/Spliced
  - Constant time access to RNA-seq reads
- Assembling unspliced reads into blocks (graph nodes)



Linking blocks with spliced reads (graph edges)



# Hashing of the input reads



# Assembling and Linking

Assembly of unspliced reads

RF=1897

cagggtaccgcgGATGATTACGTA

LF=1897

GATGATTACGTATGATTACGTAGG

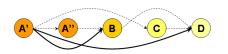
LF=5674

TGATTACGTAGGCgaatttgatac

A'cagggtaccgcggatgattacgtatgattacgtaggcgaatttgatac

Linking with spliced reads

A' cagggtaccgcggatgattacgtatgattacgtaggCGAATTTGATAC



CGAATTTGATAC GCTGGCTAGTCA

GCTGGCTAGTCAgtcatttgcat

CGAGTTTGATAC TAGGTCAGGTTT

D TAGGTCAGGTTTgtagcatgagtatta



### **Experiment on Simulated Data**

#### Data from: 112 genes used as training set in EGASP\*

- $\circ$  22.8  $\times$  10<sup>6</sup> simulated reads
- read length: 64bp
- % of mutated reads: 0, 2, 4, 8, 16

#### Results

- $\circ \sim$  40 genes "correctly reconstructed"
- 67 minutes
- Average  $S_n = 0.868$
- Average  $S_n = 0.765$



<sup>\*</sup>Guigò et al., Genome Biology, 2006

### An Example: gene L1CAM

#### Prediction summary

Predicted nodes: 22

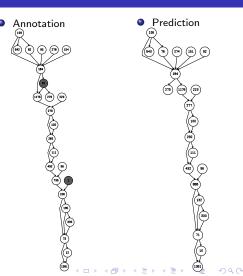
• Predicted arcs: 27

•  $S_n$  (nodes): 0.84

•  $S_p$  (nodes): 0.95

•  $S_n$  (arcs): 0.71

•  $S_p$  (arcs): 0.82



### Experiment on Real Data

#### RNA-seq data from ENCODE/Caltech

- $\circ$  2  $\times$  10<sup>9</sup> reads
- read length: 75bp (Illumina)
- unknown error

#### Results

- 210 minutes
- Average  $S_n = 0.358$
- Average  $S_p = 0.294$

#### Issues and Future Work

#### Issues

- SNP
- Read error
- Splice junctions not uniquely identified
- Some AS variations are hard to characterize

#### Future Works

- Extract AS events (exon skipping, mutually exclusive exons, etc.) from isoform graph
- Use a reference genome to predict AS variants in a donor genome (also represented with RNA-seq reads)
- Genome-wide experiment on real data from different sequencing technologies



#### **Conclusions**

- New method for AS variants inference from NGS data
- Efficient in theory and practice
- 2 k-mers/read
- No error  $\rightarrow$  Good performance
- Extremely scalable approach
- Ongoing implementation development
  - Improving performances on real data
  - SNP
  - Error correction
  - Intron/Exon refinement (involving the genome)

